

Setting Limits in CDF New Particle Searches

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Workshop on Confidence Limits

27-28 March, 2000 at Fermilab

- Introduction
- Single-channel counting experiments
- Particle searches using spectrum fits.
- Combining channels
- Remarks

Introduction

In Run I at CDF, we searched for many non-standard-model particles.

Despite our wishes, we see no clear evidences to claim a discovery of such a particle, so far. This talk focuses on simply describing how we set limits in many of these studies.

(Not intended to focus on detail discussion of mathematical techniques & their validities

note 1:

For Run II new particle search capability studies, see
<http://fnth37.fnal.gov/susy.html>,
<http://runiicomputing.fnal.gov/strongdynamics/web/strongdynamics.html>

note 2:

See John Conway's talk from Jan. CERN CL workshop for "What if there is an excess?"

note 3:

"e" "e" $\gamma\gamma E_T$ event

→ no model, so simply estimate
SM probability ... a posteriori
Phys. Rev. D 59, 092002 (1999)

Setting Limits in Single-channel Counting Experiment

We search for a new particle & figure out:

- {
 - No : # of observed events in L pb^{-1} .
 - $\mu_b \pm \sigma_{\mu_b}$: # of expected background events and its uncertainty.
 - $\mu_s \pm \sigma_{\mu_s}$: # of expected signal events (From a model) and its uncertainty.

Using these numbers, how can we constrain the theoretical model quantitatively?

- We want to use a method which
- * is reasonable and well defined.
 - * takes into account uncertainties in background
 - * takes into account uncertainties in signal acceptance.

CDF uses the 'Standard' frequentist definition of a 95% C.L. limit with a (Bayesian) integration over background and Acceptance uncertainties.

Poisson Distribution, $P(n_0 | \mu)$

$$P(n_0 | \mu) = \frac{\mu^{n_0} e^{-\mu}}{n_0!}$$

Poisson distribution with mean value μ gives probability of observing n_0 events.

Confidence Limit (C.L.) &

the Upper Limit (N_{limit})

N_{limit} : Number of expected events as that value of μ for which there is some probability ϵ to observe n_0 or fewer events.

$$\epsilon = \sum_{n=0}^{n_0} P(n | \mu)$$

then

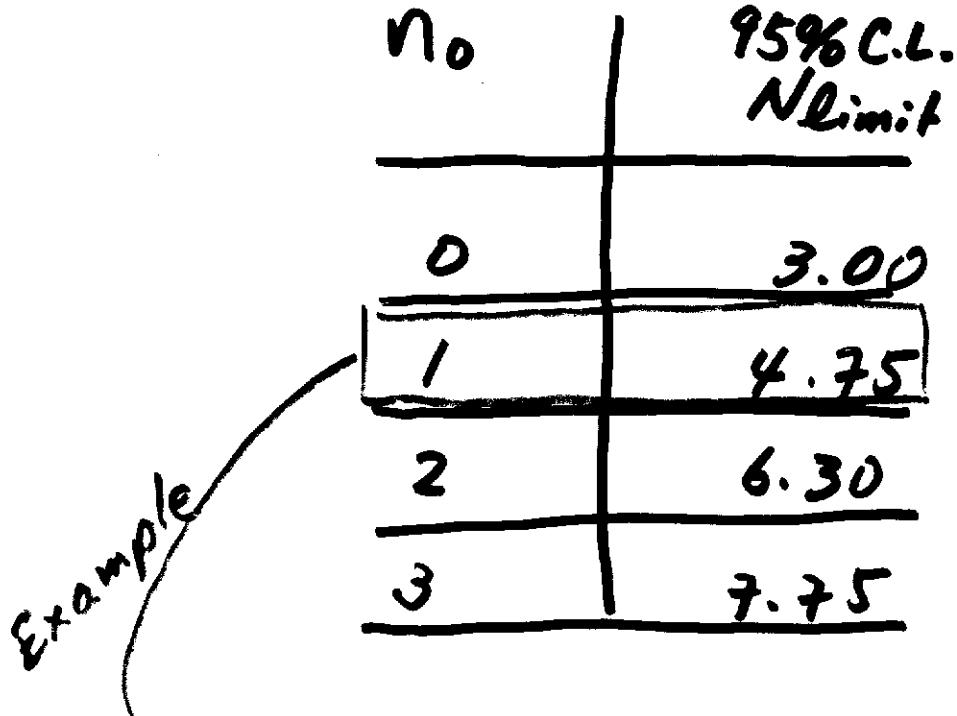
$$CL = 1 - \epsilon$$

In practice, we calculate N_{limit} by varying μ until finding the value of ϵ corresponding to the desired CL

→ N_{limit} is the resultant value of μ .

example

with no (uncertainty)
background



→ Poisson distribution with mean value = 4.75 gives 5% probability that we observe 1 or 0 events.

We exclude a model which expects more than 4.75 events at 95% C.L.

$$\rightarrow \boxed{(\sigma \cdot B_n)_{\text{limit}} = \frac{N_{\text{limit}}}{\Sigma_{\text{tot}} \cdot L}} \quad \begin{matrix} \text{cross section} \\ \text{Limit} \end{matrix}$$

note: N_{limit} is a smaller number (appears 'better') at 90% C.L. ($< 95\% \text{ C.L.}$) with the same experimental results.

Σ_{tot} : Total signal efficiency / : Inter luminosity

Setting Limits when we expect
 μ_b background events in
 n_o observed events

(no uncertainties, yet)

- n_o : # of observed events
- μ_b : expected # of background events in n_o
- n_b : # of background events in n_o ($n_b \leq n_o$)
- μ_s : mean # of signal event expected

$$\epsilon = \frac{\sum_{n=0}^{n_o} P(n | \mu_s + \mu_b)}{\sum_{n=0}^{n_o} P(n | \mu_b)}$$

$$= \frac{\sum_{n=0}^{n_o} \frac{(\mu_s + \mu_b)^n \cdot e^{-(\mu_s + \mu_b)}}{n!}}{\sum_{n=0}^{n_o} \frac{\mu_b^n \cdot e^{-\mu_b}}{n!}}$$

N_{limit} (for the signal) at C.L. = ϵ represents the value of μ_s for which there is some probability $1 - \epsilon$ to observe more than n_o events
 $(= \epsilon \text{ prob. to observe } n_o \text{ or less events})$

and have $n_b \leq n_o$.

note: For a given No (observed events), we get "better" limit for higher # of expected background μ_b .
(downward fluctuation "Problem")

example:

Suppose we observe 1 event, when

- 1) expecting 1 background event
- 2) expecting 10 background events.

→ 2) → Gives 'better' limit!

→ Suggestion: Check ~~what's~~'s

Incorporating Uncertainties

To incorporate

$$\left\{ \begin{array}{l} \sigma_{\mu_b} : \text{uncertainty in background} \\ \sigma_{\mu_s} : \text{uncertainty in signal} \end{array} \right.$$

we integrate over all possible values of the true value μ_s and μ_b .

This is a Bayesian technique, and assumes a Gaussian prior probability-density function for the true values μ_s and μ_b .

N_{limit} at $1-\epsilon$ C.L. is equal to μ_s

which satisfies:

$$\epsilon = \frac{\sum_{n=0}^{\infty} \frac{1}{\sqrt{2\pi\sigma_{\mu_s}^2}} \int_0^\infty \int_0^\infty P(n|\mu_s + \mu_b) e^{-\frac{(\mu - \mu_s)^2}{2\sigma_b^2}} e^{-\frac{(\mu_s - \mu_b)^2}{2\sigma_b^2}} d\mu d\mu_b}{\sum_{n=0}^{\infty} \int_0^\infty P(n|\mu_b) e^{-\frac{(\mu_b - \mu_b)^2}{2\sigma_b^2}} d\mu_b}$$

$$\sigma_{\mu_s} = \mu_s \cdot \frac{\sigma_{\text{tot}}}{\epsilon_{\text{tot}}} , \quad \epsilon_{\text{tot}} \leftarrow \text{total efficiency for the}$$

To obtain N_{limit} at $(1-\epsilon)$ c.l.

One could perform the formidable integral
(in previous page).

Instead, we solve this by performing
a large ensemble of random pseudo-experiments
varying the expected number of signal and
background about their nominal values
according to a gaussian distribution.

→ standardized in program poilim.f

(J. Conway)
checked by various
people.

Single-channel Counting Experiment

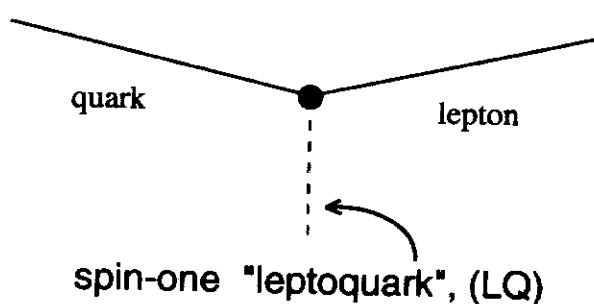
→ Used in many CDF new particle searches
SUSY, Charged Higgs, Leptoquarks
etc ...

example → **Pati-Salam Model and "leptoquark" Search**

Pati-Salam Model: Simple gauge theory of quark-lepton unification
Lepton number as the Fourth "color"

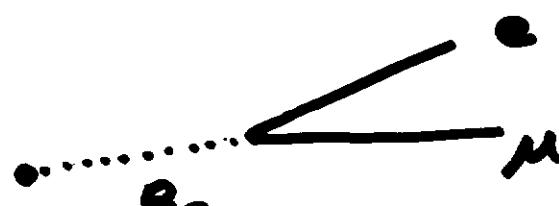
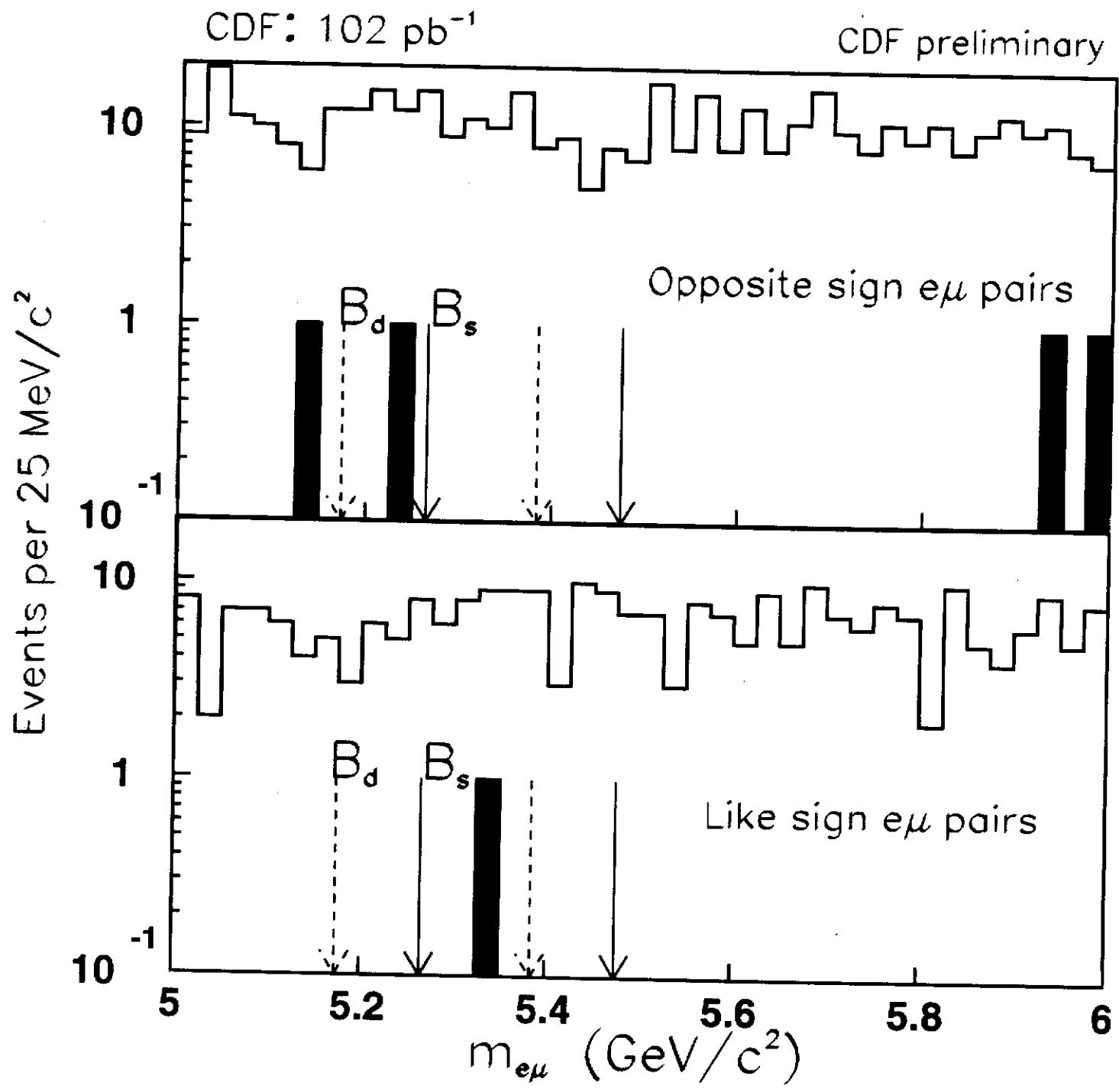
Symmetry breaking: $SU(4)_C \rightarrow SU(3)_C$ at high E

the theory predicts: Heavy Pati-Salam Bosons (leptoquarks)



PRL Dec. 28, 1998

Vol 81, 26 pp. 5742 - 5747



N^{limit}

The limit is dominated by Poisson statistic. The Poisson limits are:

- $N^{limit} = 2.3$ (3.0) events for observing zero events (B_s case).
- $N^{limit} = 3.9$ (4.7) events for observing one events (B_d case).

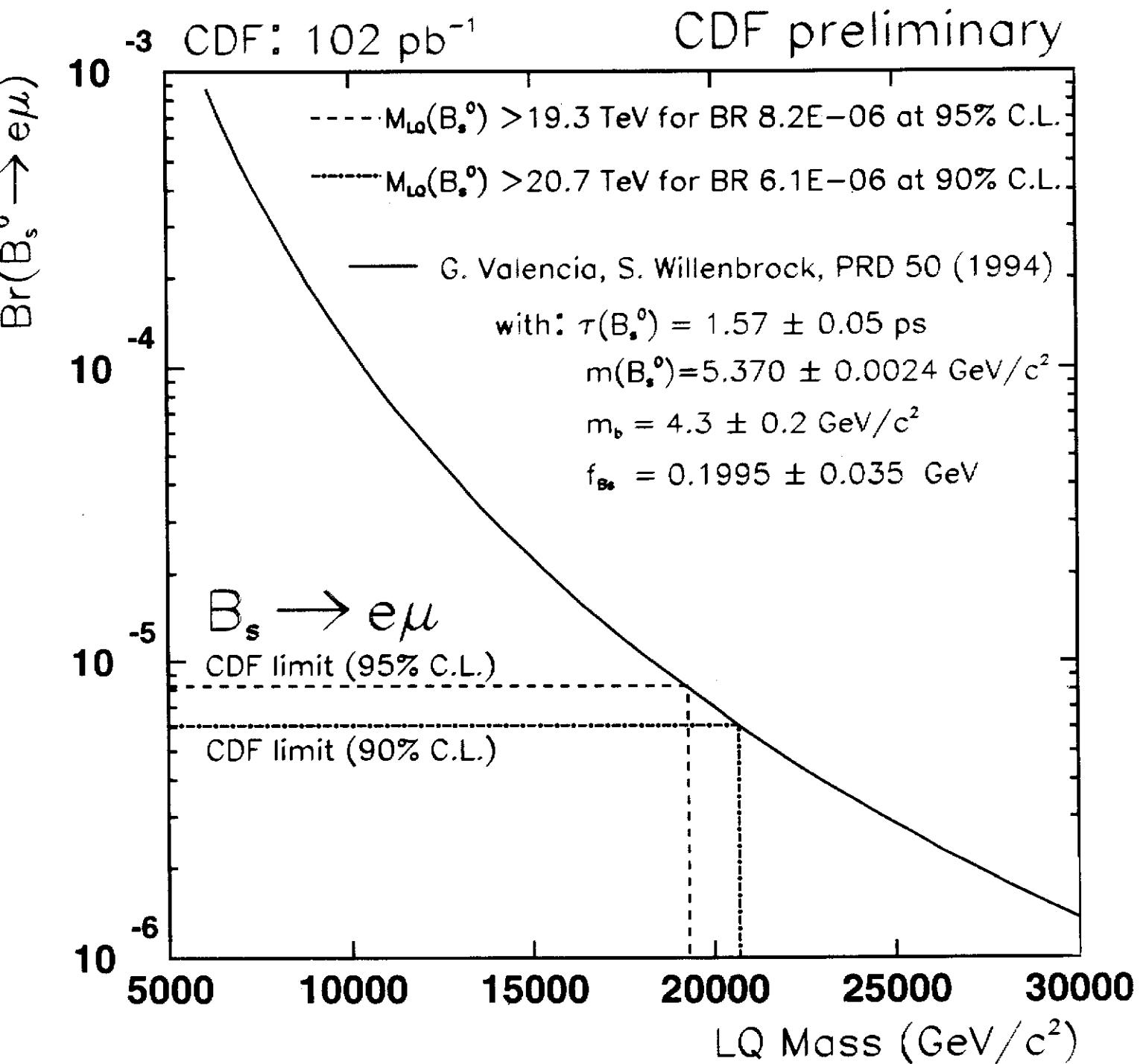
(no uncertainties)
We combine the systematic uncertainties with the poisson limits using the convoluted-likelihood function method.

uncertainty on B-meson cross section	$\pm 23\%$
uncertainty on acceptance and cut efficiencies	$\pm 10\%$
uncertainty on $\int \mathcal{L} dt$	$\pm 6.9\%$
Total	$\pm 26.0\%$

Table 1: Systematic errors in %

- (with uncertainties)*
- $N^{limit} = 2.52$ (3.38) events for observing zero events (B_s case).
 - $N^{limit} = 4.34$ (5.52) events for observing one events (B_d case).

\uparrow \uparrow 95% C.L.
90% C.L.



Advantages

- Simple extension of standard frequentist method → easy to understand
- No issue of choice of Prior Prob. density function.
- Can be easily extended to take into account correlated uncertainties.
- Can always quote an upper limit even if there is an excess (no flip-flopping)

Disadvantages

- Suffers from downward fluctuation "problem".
- How do we combine channels ?
- Only works for single-channel counting experiment, not fit to spectra.

Limits from Spectra and Combining Channels

In searching for new particles using a fit to spectra, CDF typically uses binned likelihood fit method, allowing the signal cross section multiplier, f , to float.

$$\mu_i = \mu_{b_i} + f \cdot \mu_{s_i}$$

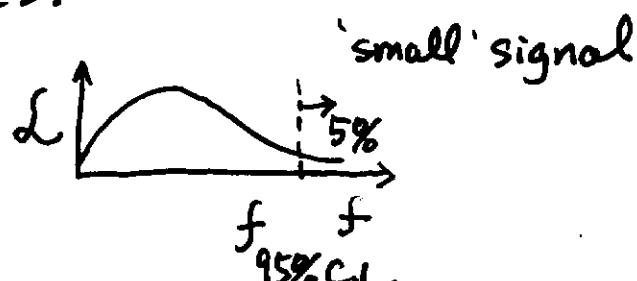
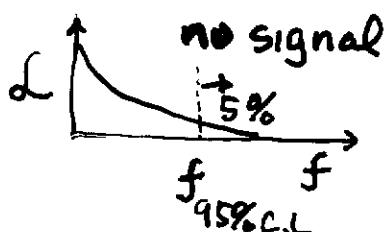
Likelihood function

$$L = (\text{Norm fac}) * \prod_{i=\text{bin}}^{\text{Nbins}} \frac{\mu_i^{n_i} \cdot e^{-\mu_i}}{n_i!}$$

$$\text{Norm fac} \leftarrow 1.0 = \int_0^{\infty} L(f) df$$

To Obtain 95% C.L. Limit

Plot the likelihood function $L(f)$, then find value of f beyond which 5% of the total integral of the L lies.



If $f_{95\% \text{ C.L.}} < 1$ then the theoretical prediction is not valid at (at least) 95% C.L.

- Uncertainties and signal & background are taking into account by integrating over them in the same way as described for single channel counting experiment
- Multiple channel results can be combined by multiplying the likelihoods for the different channel/s.

In a case,

Single channel, single bin, no uncertainty

\hookrightarrow equivalent to the frequentist method

Analyses used this techniques

many

for example,

* Z' search (ee) & ($\mu\mu$) channels

* TC (π_T, p_T) search, (leptonic)+(Hadronic)

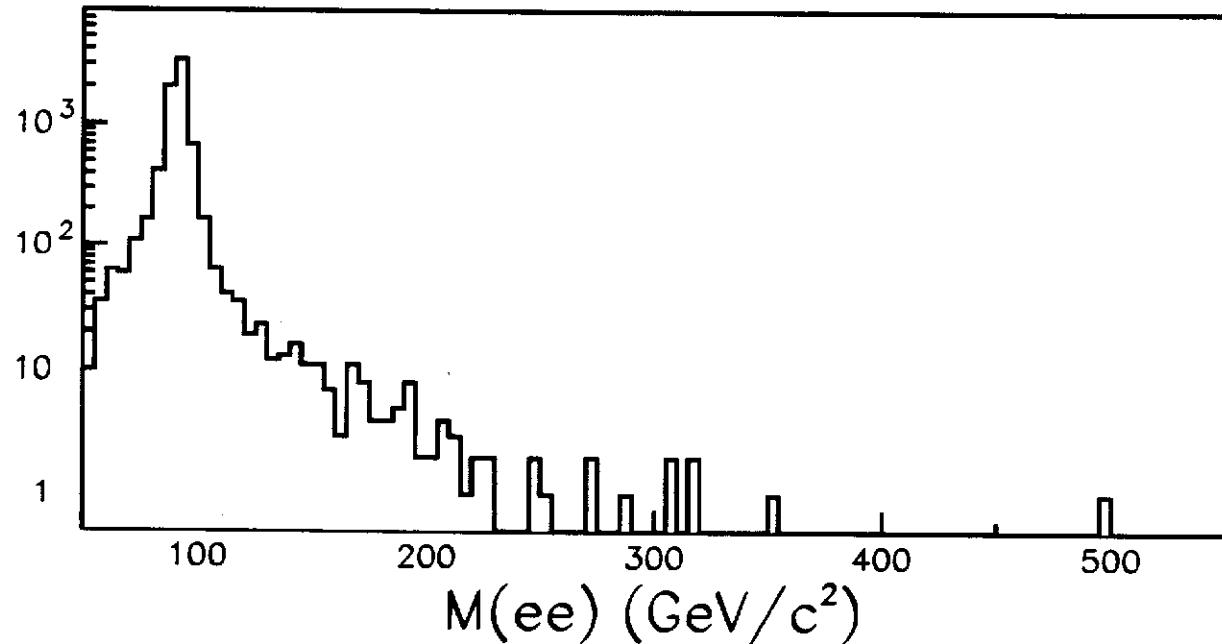
* SM. H_0 search, (leptonic)+(Hadronic)

* 4th-gen. $b' \rightarrow bZ$ search

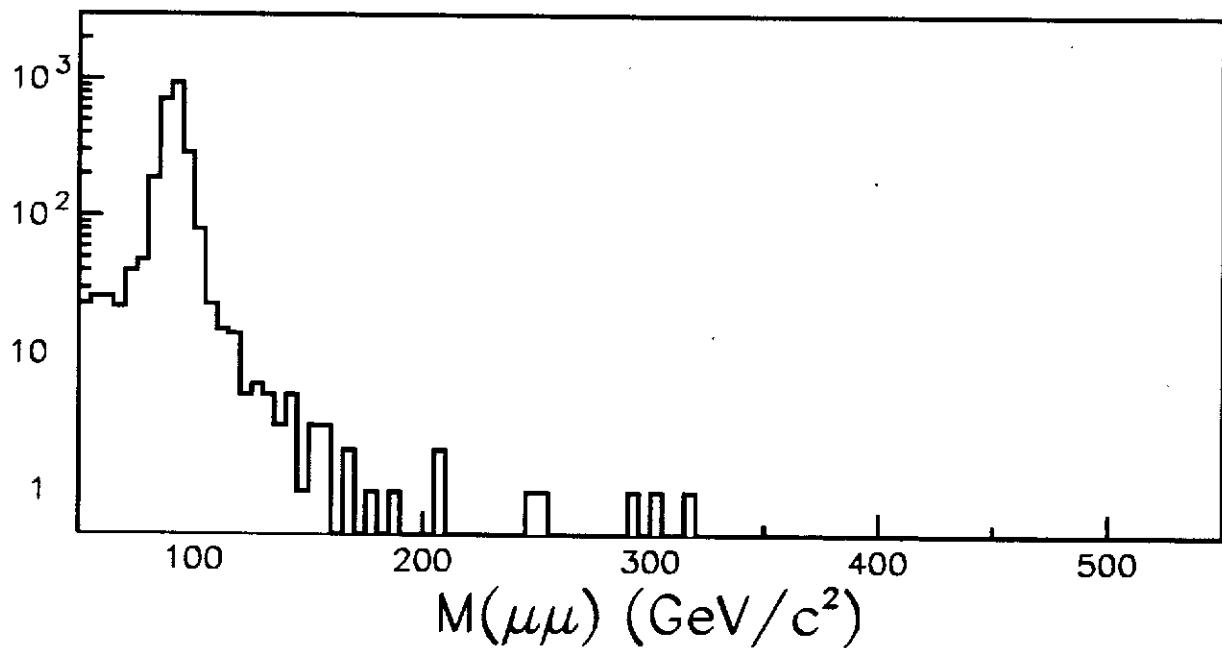
--- etc.

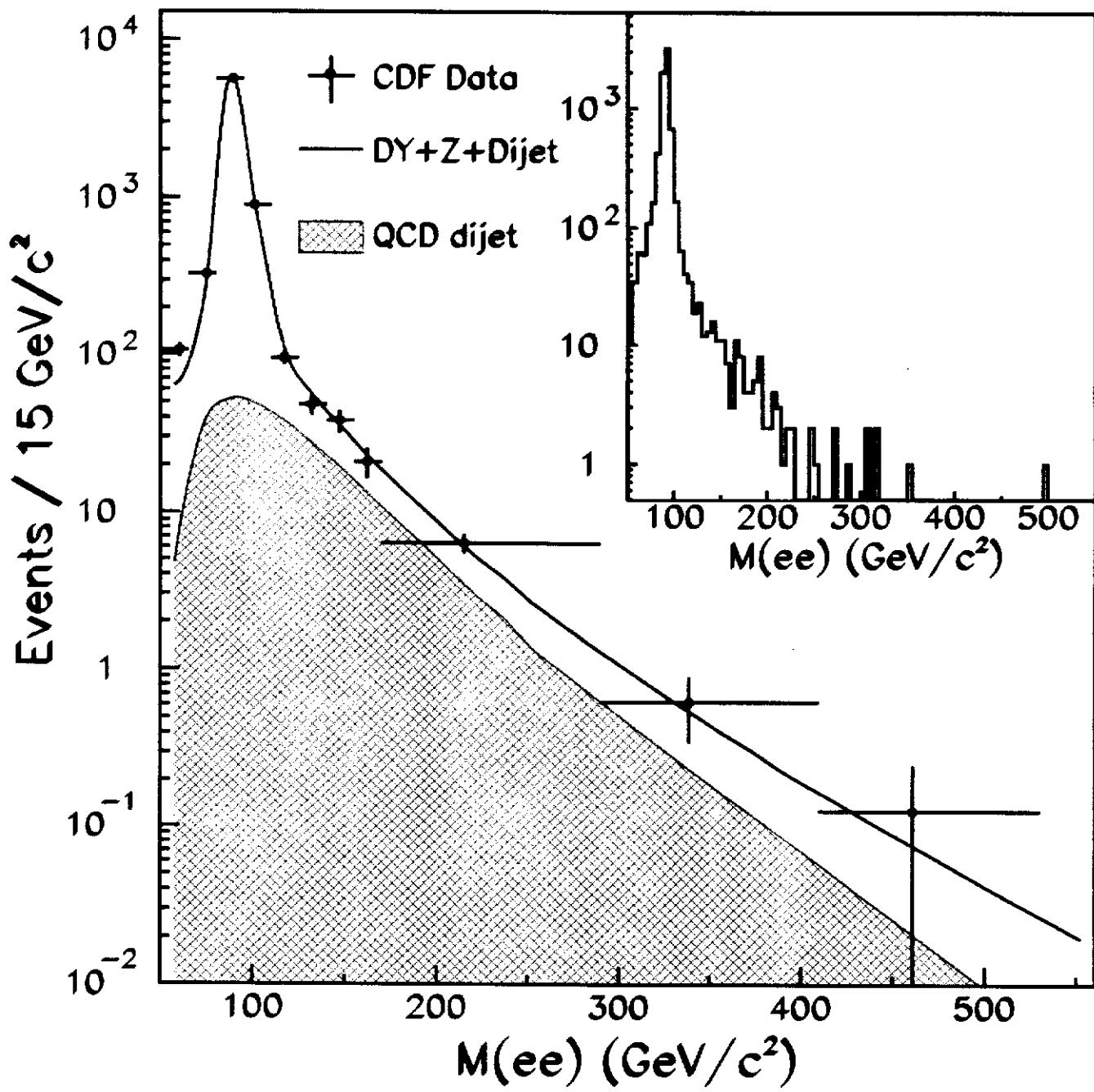
χ' Search

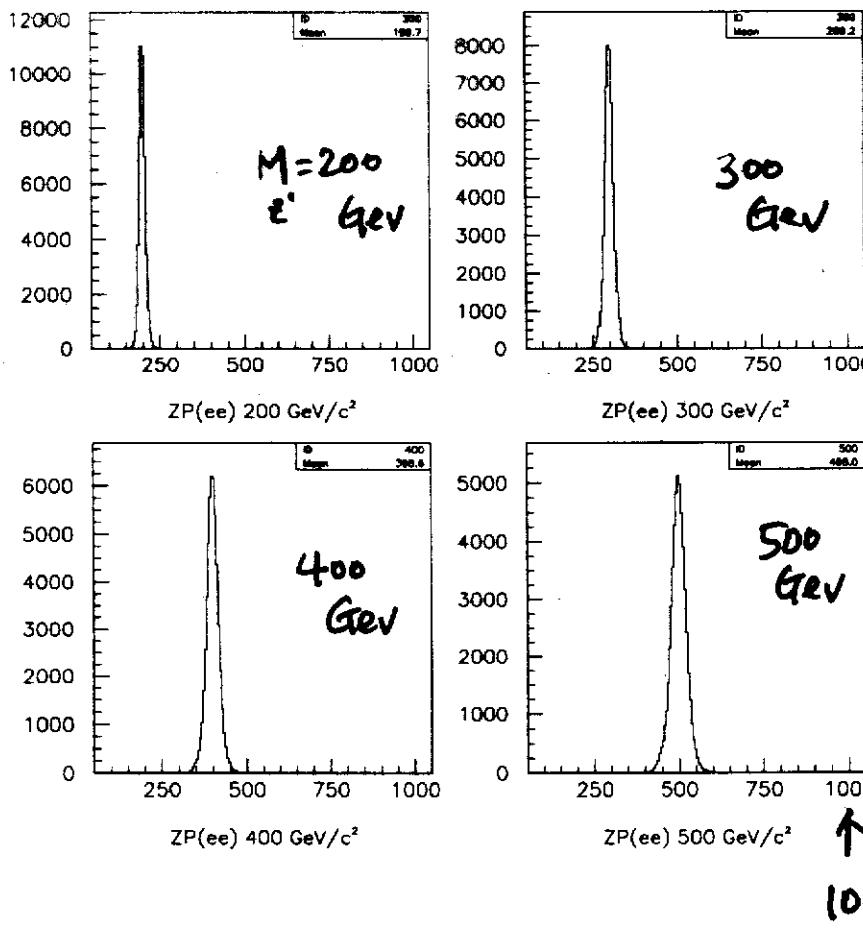
Events / 5 GeV/ c^2



Events / 5 GeV/ c^2







signal
 $M(Z' \rightarrow ee)$
simulation

Fig. 3. The figure shows distributions of Z' events in dielectron invariant mass for events after the detector smearing effects are accounted for. The distributions are shown for a Standard model Z' of mass 200, 300, 400 and 500 GeV respectively.

$$Lik(\mu_i) = (Normfac) \times \prod_{i=bins} \frac{\mu_i^N}{N!} e^{-\mu_i} \quad (3)$$

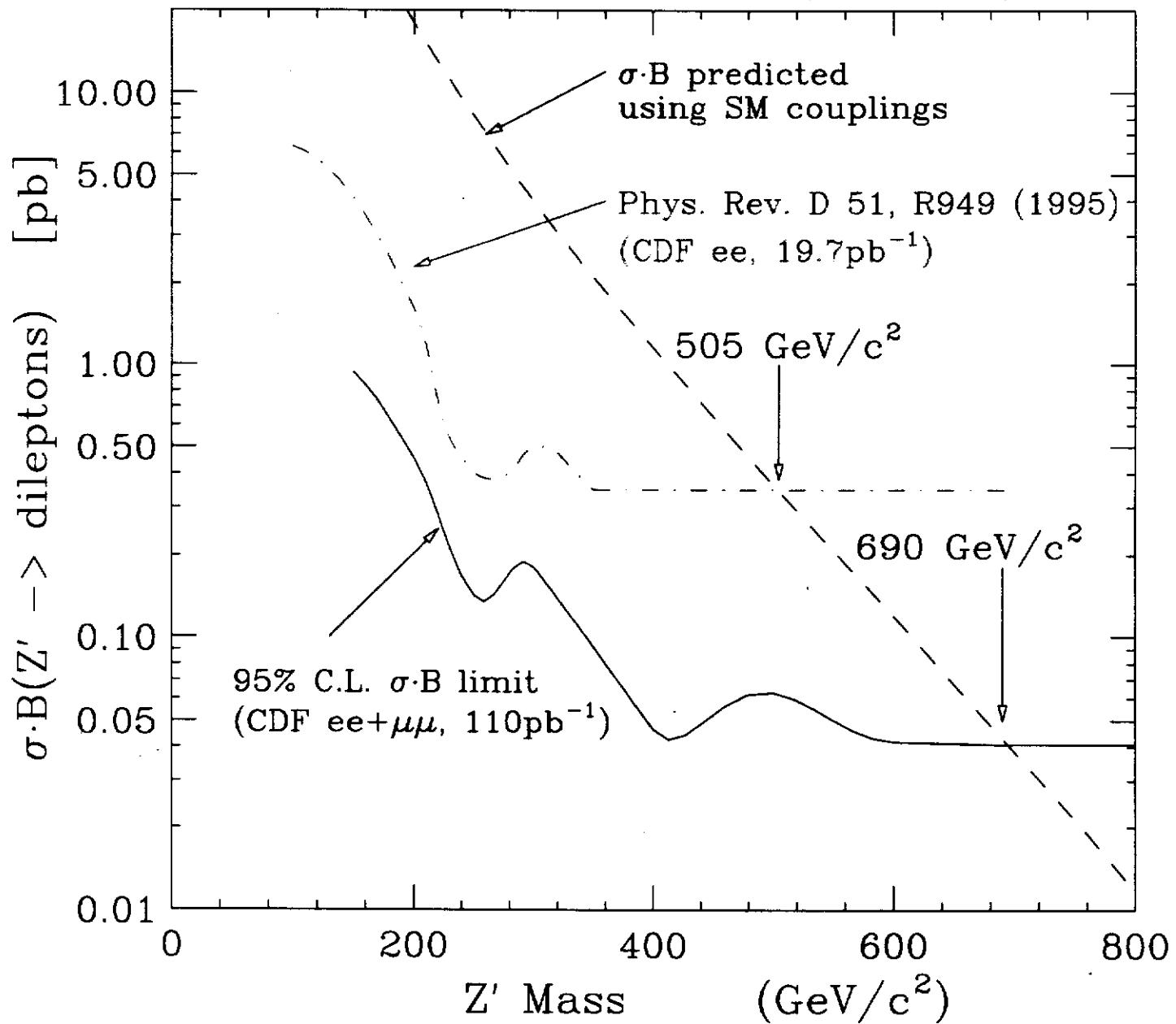
where the product over the index i runs over different mass bins and $Normfac$ is defined below. This likelihood function is just a sum over products of likelihood in each bin.

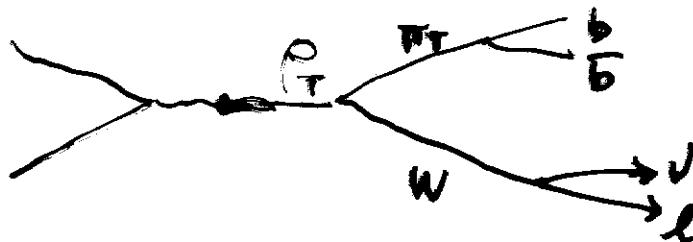
The quantity μ_i is given by :

$$\mu_i = ZDY_i + \alpha \times Z'_i \quad (4)$$

Since the μ_i are constrained by the equation 4 the quantities μ_i are no longer independent, and the likelihood is a function of a single parameter α . The quantity α is the level of the Z' content in the data with respect to the prediction from the model. The

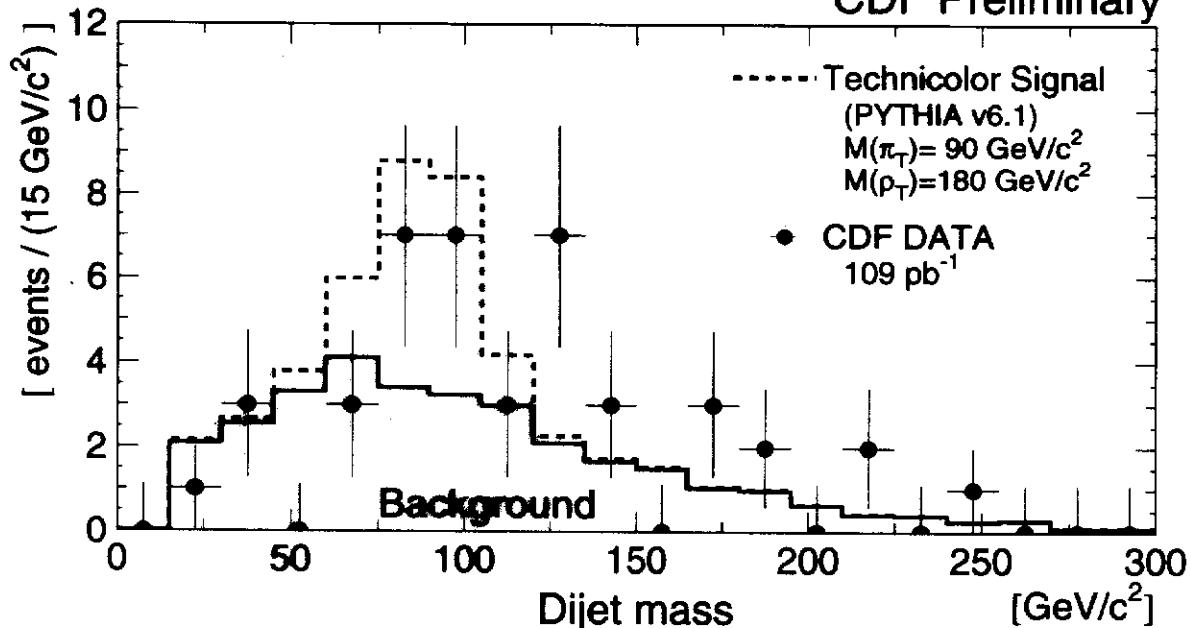
Limits on Z' production(95% C.L.)



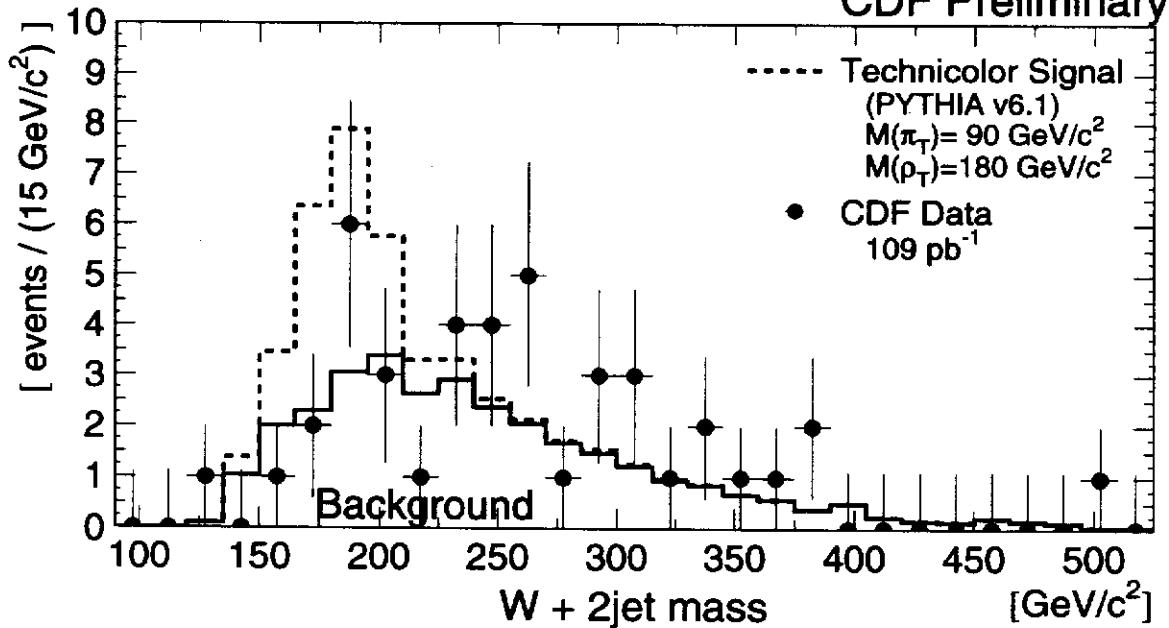


W + 2 jet with SVX b-tag

CDF Preliminary

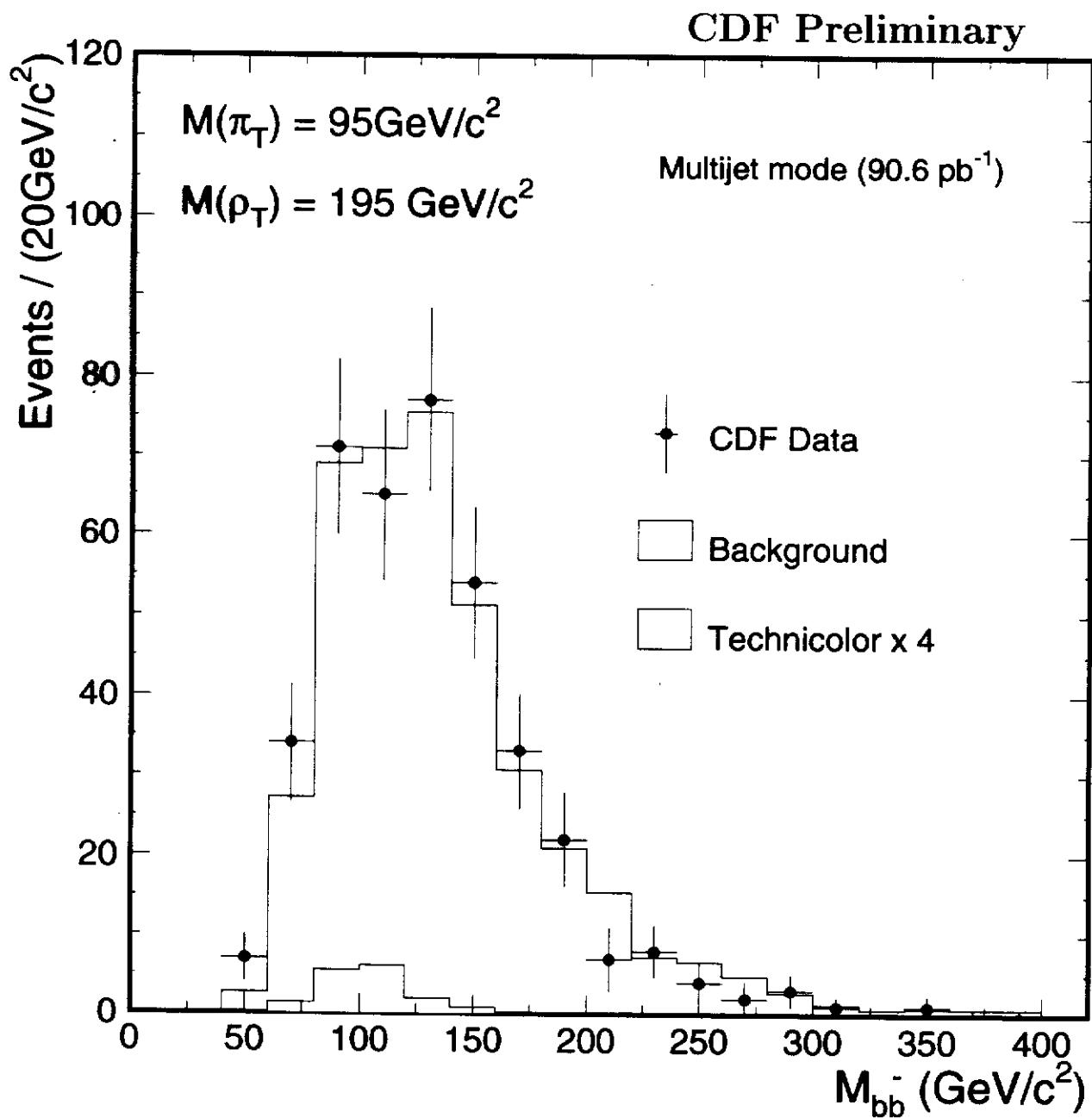


CDF Preliminary



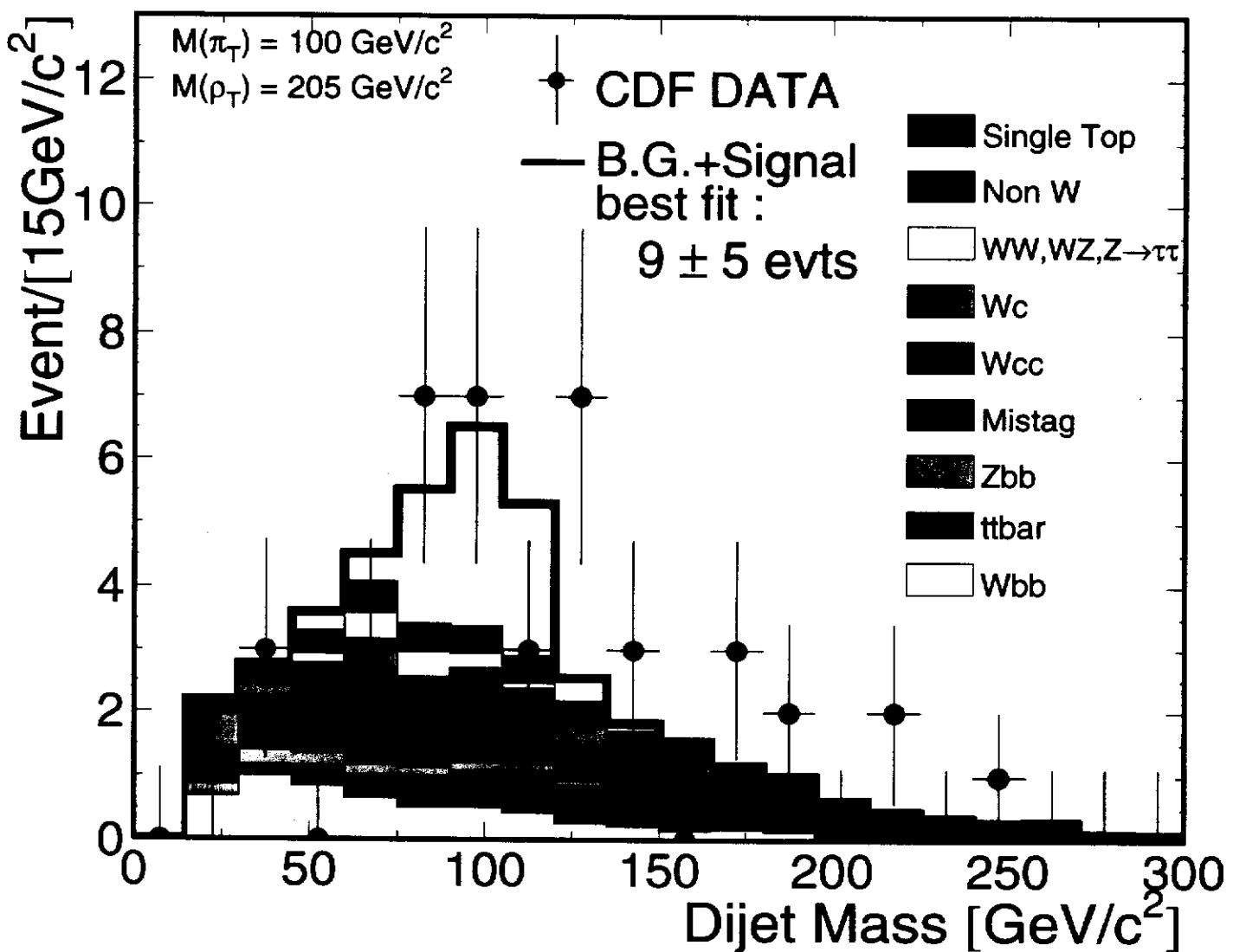
Hadronic channel

4 jets + 2 btags

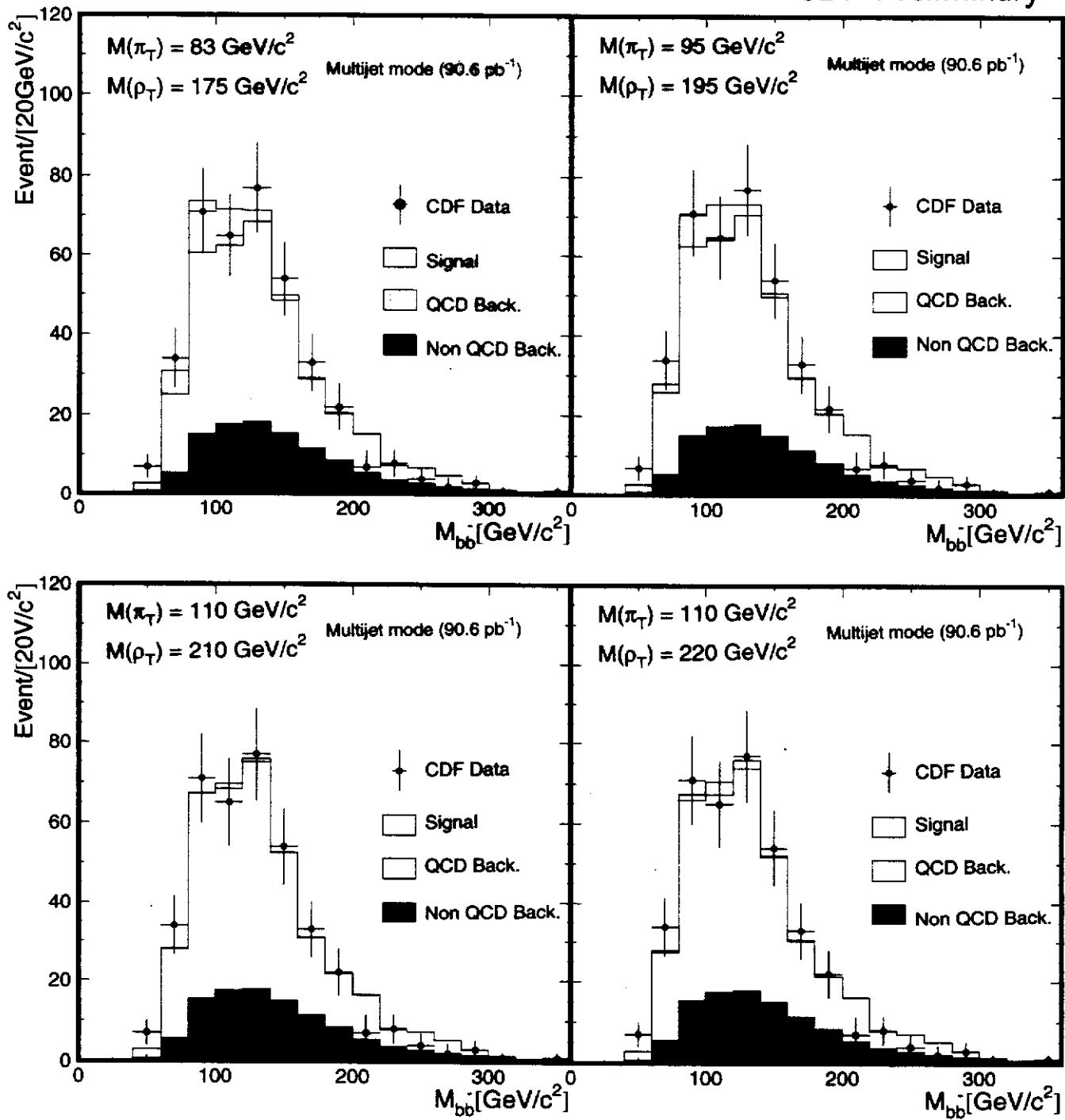


Fitting Result

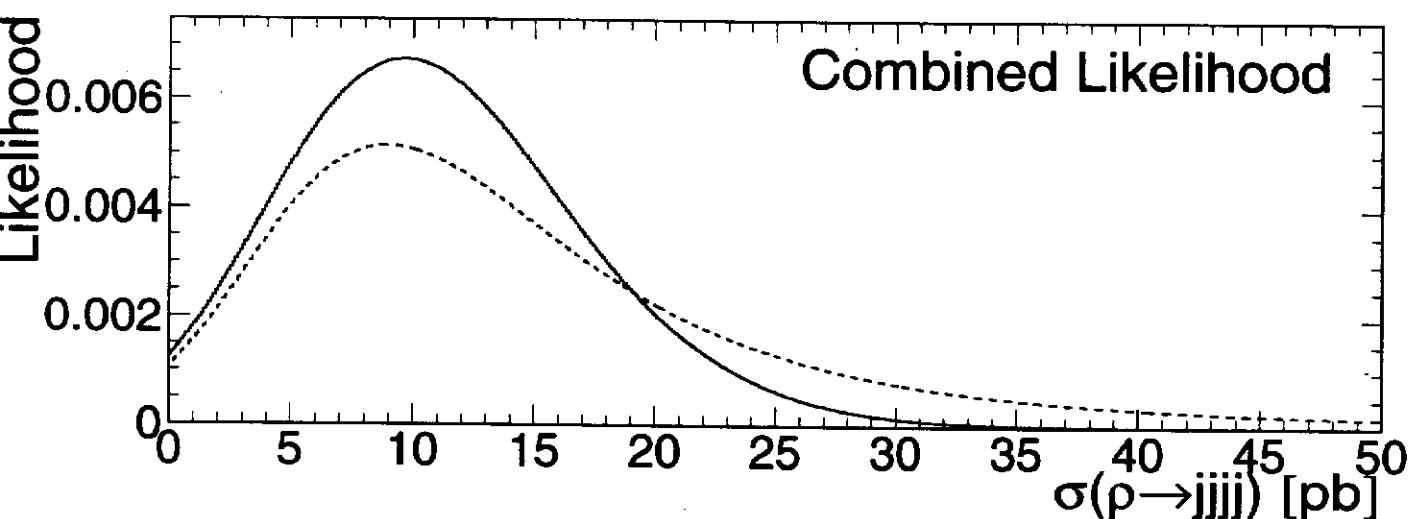
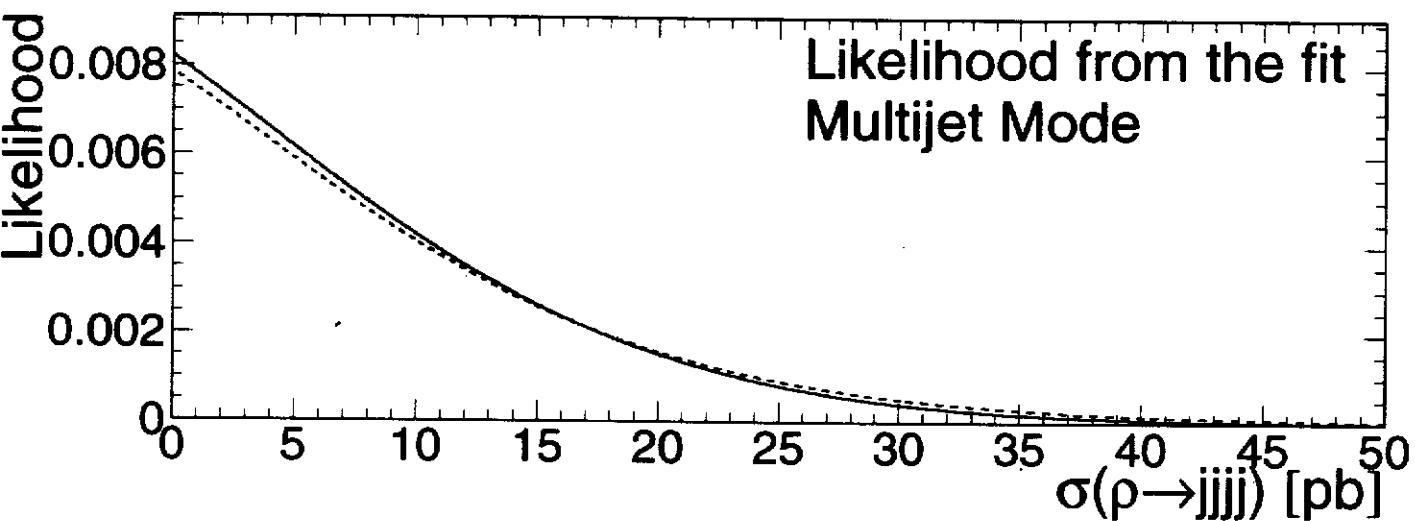
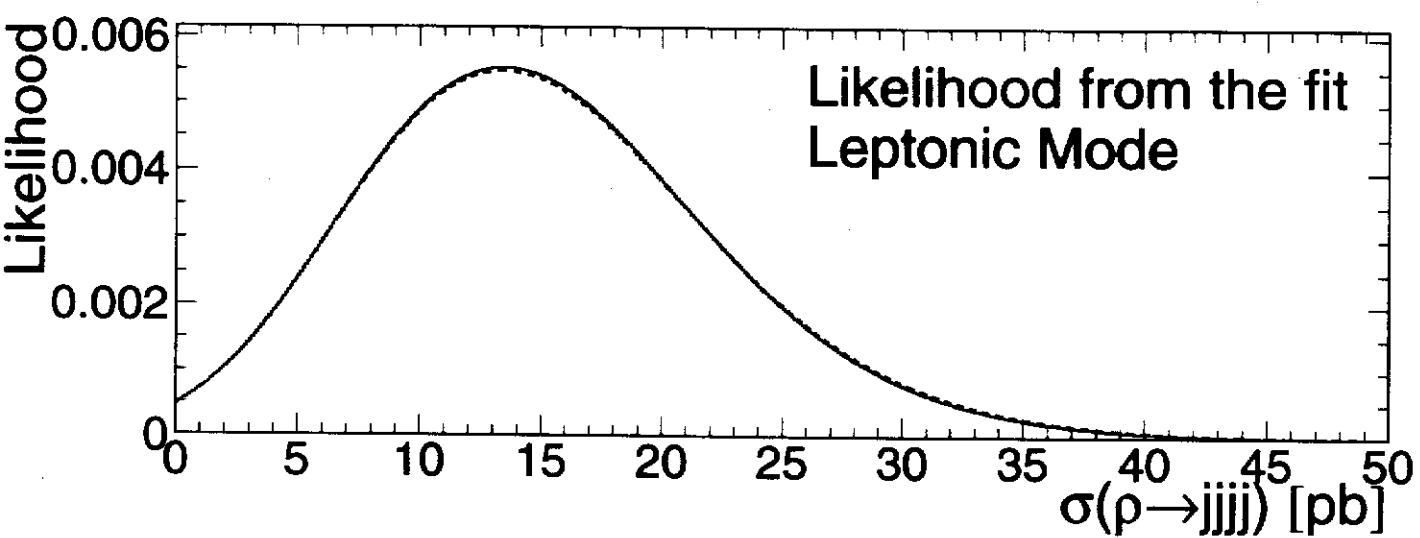
CDF Preliminary



CDF Preliminary



CDF Preliminary



Remarks

Expected Limit & Limit we actually set

Using pseudo-experiments with already known quantities fed in (no., " σ_{ub} ", " μ_b ", etc), we can estimate a range of limit (prob. of) we expect to obtain.

⇒ example, TC (hadronic) search.

In this case, we found that the final limits we obtained were consistent with what we expect.

But that is not always the case in some analyses.

Often the problems which cause significant differences in results are not due to different statistical techniques we use.

more likely, wrong ϵ_{tot} estimate, wrong background estimate, etc, etc ...